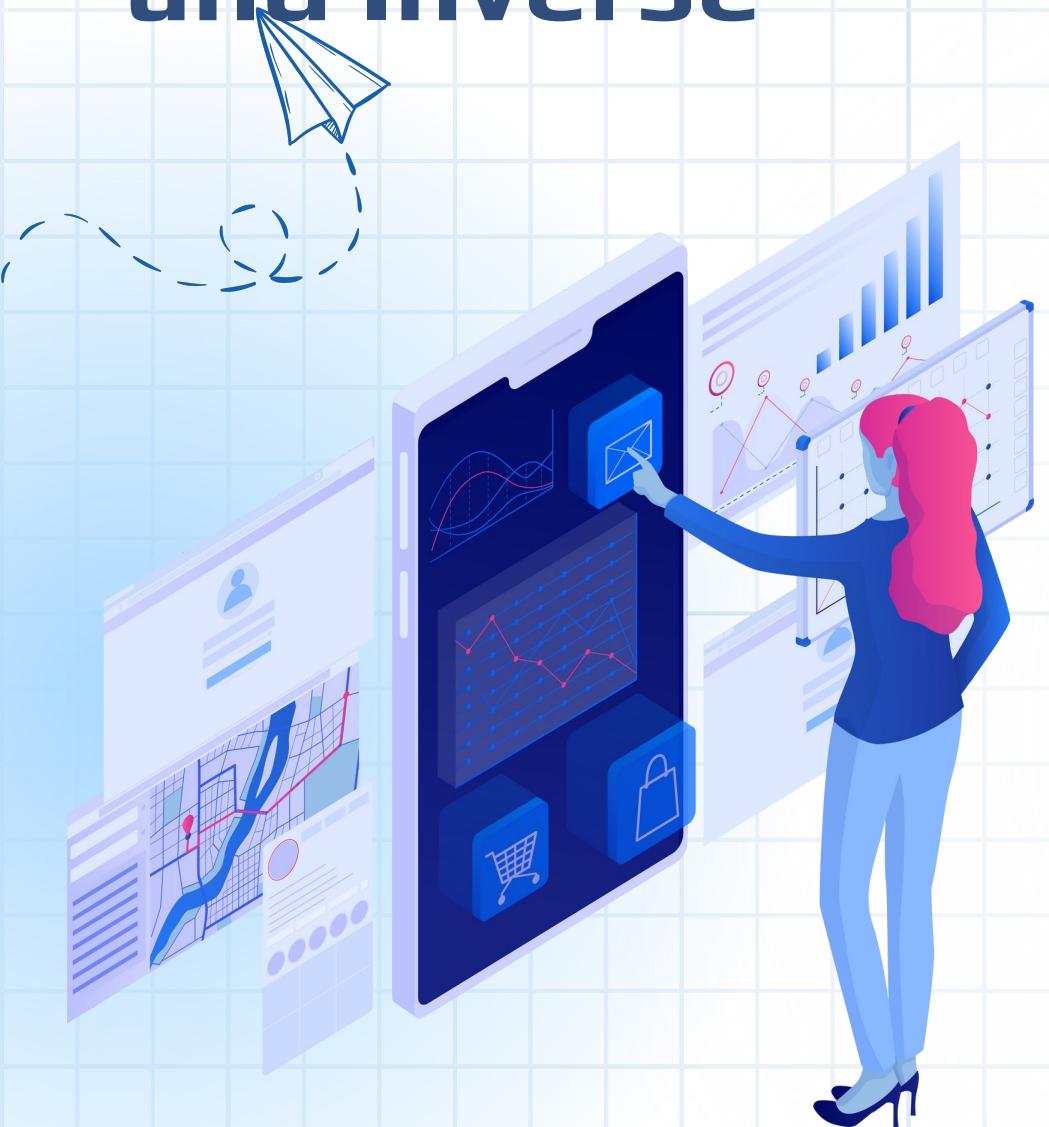


BS./BSc.

in Applied AI and Data  
Science

Linear algebra and  
numerical analysis

# Module 2.2 Symmetry, Determinants, and Inverse



- 1 **Symmetric and Square matrix**
- 2 **The determinant of a matrix**
- 3 **Inverse of a matrix**
- 4 **Rank of a matrix**



# Symmetric matrix operations in daily life

Symmetric matrices have many special properties that make them great to work with in numerous real-world applications.

Please see Examples below:

**Image Processing:** Covariance matrices, which are always symmetric, describe the relationships between image pixel values.

**Covariance Matrices:** In statistics and machine learning, symmetric covariance matrices are used to describe the relationships between different variables in a dataset. They are fundamental in techniques like principal component analysis (PCA).

**Kernel Methods:** Kernel matrices, which are often symmetric, are used in support vector machines (SVMs) and other kernel-based learning algorithms. These matrices capture the similarity between data points.

**Network Analysis:** Adjacency matrices representing undirected graphs are symmetric. These matrices analyze social networks, web graphs, and other interconnected systems.



# What does it mean for a matrix to be symmetric?

- ❖ It means that the corresponding rows and columns are equal. And that means that nothing happens to the matrix when you swap the rows and columns.

A Square matrix is symmetric if it is equal to its transpose:

$$A = A^T \quad A = \begin{bmatrix} a & b \\ b & d \end{bmatrix}$$

$$A^T = \begin{bmatrix} a & b \\ b & d \end{bmatrix}$$

# Creating Symmetric Matrices from Nonsymmetric Matrices



This may be surprising at first, but multiplying any matrix—even a nonsquare and nonsymmetric matrix—by its transpose will produce a square symmetric matrix.

The proof is obtained simply by considering the matrix sizes: if  $A$  is  $M \times N$ , then  $A^T A$  is  $(N \times M) (M \times N)$ , which means the product matrix is of size  $N \times N$ . You can work through the same logic for  $AA^T$ .

Now, to prove symmetry. Recall that the definition of a symmetric matrix equals its transpose. So, let's transpose  $A^T A$ , do some algebra, and see what happens. Make sure you can follow each step here; the proof relies on the **LIVE EVIL** rule:



# Symmetric Matrix operation in Python

In this exercise, you will write a Python function that checks whether a matrix is symmetric. **Here's how you can do it:**

**It should take a matrix as input and output a boolean True if the matrix is symmetric or False if the matrix is nonsymmetric.**

# Code for symmetric matrix creation



- **create\_symmetric\_matrix(size):** It initializes a square matrix of the given size with zeros using `np.zeros()`.
  - It iterates through the upper triangle of the matrix (including the main diagonal) using nested loops.
  - Inside the loops, it generates a random integer using `np.random.randint()` and assigns it to the current element `matrix[i, j]`.
  - To ensure symmetry, it also assigns the same value to the corresponding element `matrix[j, i]`.
  - It returns the generated symmetric matrix.
- **is\_symmetric(matrix):** It takes a NumPy array as input.
- It uses `np.array_equal()` to compare the matrix with its transpose (`matrix.T`).
  - It returns `True` if the matrix is symmetric (equal to its transpose), and `False` otherwise.

# Question

Show that the Product of The following Matrix and Its Transpose Is Symmetric

Let  $A$  be the  $2 \times 3$  matrix



$$A = \begin{bmatrix} 1 & -2 & 4 \\ 3 & 0 & -5 \end{bmatrix}$$



# Solution

$$A^T A = \begin{bmatrix} 1 & 3 \\ -2 & 0 \\ 4 & -5 \end{bmatrix} \begin{bmatrix} 1 & -2 & 4 \\ 3 & 0 & -5 \end{bmatrix} = \begin{bmatrix} 10 & -2 & -11 \\ -2 & 4 & -8 \\ -11 & -8 & 41 \end{bmatrix}$$

$$AA^T = \begin{bmatrix} 1 & -2 & 4 \\ 3 & 0 & -5 \end{bmatrix} \begin{bmatrix} 1 & 3 \\ -2 & 0 \\ 4 & -5 \end{bmatrix} = \begin{bmatrix} 21 & -17 \\ -17 & 34 \end{bmatrix}$$

Observe that  $A^T A$  and  $AA^T$  are symmetric as expected.



# The inverse of a Matrix and real-world applications

**Solving Linear Systems:** A fundamental use of matrix inverses is to solve systems of linear equations.

In **data engineering**, this is relevant when dealing with linear models that represent relationships between variables. For example, in certain types of statistical modeling, you might need to solve for unknown parameters in a linear system.

**Computational Cost:** Calculating the inverse of large matrices can be computationally expensive. In data engineering, where datasets can be massive, alternative methods for solving linear systems (like decomposition techniques) are often preferred.



# Computing inverse of a Matrix

The determinant is required to compute the inverse of a matrix. Each **square matrix A** has a unit scalar value called the determinant of A, denoted by **det A or |A|**

If  $A = \begin{bmatrix} 1 & 2 \\ 6 & 5 \end{bmatrix}$

then  $|A| = \begin{vmatrix} 1 & 2 \\ 6 & 5 \end{vmatrix}$



# Properties of inverse Matrix

## Example

If  $\mathbf{A} = [A]$  is a single element ( $1 \times 1$ ),  
then the determinant is defined as the  
value of the element

Then  $|A| = \det A = a_{11}$

If  $\mathbf{A}$  is  $(n \times n)$ , its determinant may be  
defined in terms of order  $(n-1)$  or less.



# Challenges in determining determinants

Once you get to **4x4 matrices**, the determinant becomes a hassle to compute unless the matrix has many carefully placed zeros. But I know you're curious, so the Equation below shows the **determinant of a 4x4 matrix**. Yeah, good luck with that

$$\begin{vmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{vmatrix} = afkp - aflo - agjp + agln + ahjo - ahkn - bekp + belo + bgip - bglm - bhio + bhkm + cejp - celn - cfip + cfim + chin - chjm - dejo + dekn + dfio - dfkm - dgin + dgjm$$

Anyway, the point is that if you ever need to compute the determinant, you use `np.linalg.det()` or `scipy.linalg.det()`.

# What is the determinant? What does it mean and how do we interpret it?



**Computing the determinant is time-consuming and tedious.**

**Example: Computing the determinant of a 2x2 matrix**

$$\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$$

You can see from this equation that the determinant is not limited to integers or positive values. Depending on the numerical values in the matrix, the determinant can be **-1223729358** or **+0.0000002** or any other number. (The determinant will always be a real number for a real-valued matrix.)

# Question



**Find the determinant of a  $3 \times 3$  matrix given below**

Find the determinant of the matrix  $\begin{pmatrix} 1 & 2 & 1 \\ 0 & 3 & 4 \\ 3 & 1 & 4 \end{pmatrix}$ .



# Solution



$$\begin{aligned}|\mathbf{A}| &= 1 \begin{vmatrix} 3 & 4 \\ 1 & 4 \end{vmatrix} - 2 \begin{vmatrix} 0 & 4 \\ 3 & 4 \end{vmatrix} + 1 \begin{vmatrix} 0 & 3 \\ 3 & 1 \end{vmatrix} \\&= 1(3 \times 4 - 4 \times 1) - 2(0 \times 4 - 4 \times 3) + 1(0 \times 1 - 3 \times 3) \\&= 1(12 - 4) - 2(0 - 12) + 1(0 - 9) \\&= 8 + 24 - 9 \\&= 23\end{aligned}$$



# Recap



- **Square Matrices:** Matrices with equal rows and columns ( $n \times n$ ).
- Essential for operations like finding determinants and inverses.
- **Symmetric Matrices:** Square matrices equal to their transpose ( $A = A^T$ ).
- Elements are mirrored across the main diagonal.
- Have real eigenvalues and are important in various applications.
- All symmetric matrixes are square matrixes.
- **Determinants:** Scalar values calculated only for square matrices.
- Indicate invertibility (non-zero determinant) and are used in various calculations.
- The determinant is a property of a square matrix.

# Coming up next.....



## **Matrix Algebra: Inverse and Rank of a Matrix and Linear independence**



# Thank you

